

Introduction to Artificial Intelligence

State Space Search

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Contents

- Intro
- Theory
 - State Space Search
 - Blind State Space Search (3 algorithms)
 - Heuristic State Space Search (2 algorithms)
- Example

Most of the material used (except the examples) comes from
“The Handbook of Artificial Intelligence – Volume I”
(Avron Barr & Edward A Feigenbaum)

What

- Part of Computer Science concerned with designing intelligent computer systems
- Systems exhibiting the characteristics we associate with intelligence in human behaviour

Areas

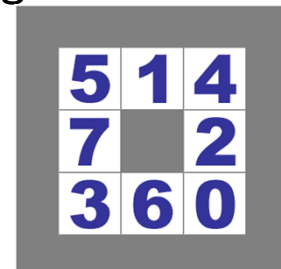
- *The areas are not distinct – most are interrelated*
- Problem Solving
 - Puzzles
 - Play games, eg chess
 - Symbolic integration of mathematical formulas
 - Some programs can improve their performance with experience
- Logical reasoning
 - Prove assertions (theorems) by manipulating a database of facts
- Language
 - Understanding of natural language
 - Translation
 - Eg Spelling checker
- Programming
 - Write computer programs based on a description

Areas – cont'd

- Learning
 - Learning from examples
- Expertise (aka Expert Systems)
 - User interacts with an Expert System via a dialogue
 - Expert feeds knowledge
- Robotics and vision
 - Manipulate robot devices (mostly in industrial applications to perform repetitive tasks)
 - Recognize objects and shadows in visual scenes
- Systems and languages
 - Time-sharing, list processing, and interactive debugging were developed in the AI research

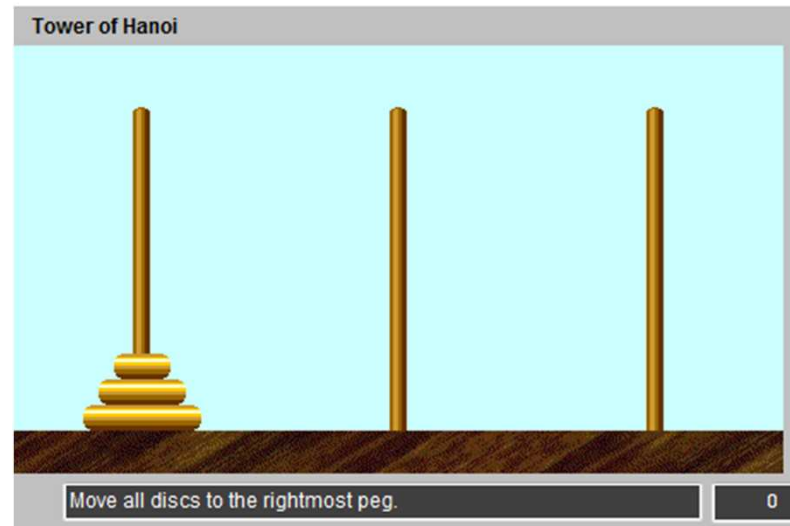
Search

- Components of search systems
 - Database : describes the current task domain and the goal
 - Set of operators: transform one state to another
 - Eg in the 8 puzzle: UP, DOWN, LEFT, RIGHT
 - Control strategy : decides what to do next
- Definition
 - Find a finite sequence of operators transforming the initial state to a goal state
- Reasoning
 - Forward : Transform original state to a goal state
 - Backward: Transform a goal state to the original state



Search

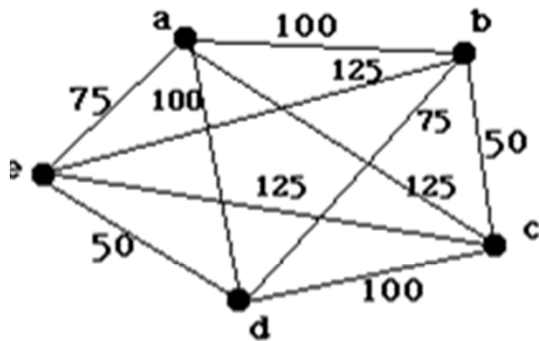
- State Space and Problem Reduction
 - State space
 - An operator produces exactly one new state
 - Problem reduction
 - An operator produces a set of subproblems, each of which have to be solved
 - Eg
 - Tower of Hanoi
 - Integrate $(f(x) + g(x)) dx$
 - » Integrate $f(x) dx$
 - » Integrate $g(x) dx$
 - » Add the results



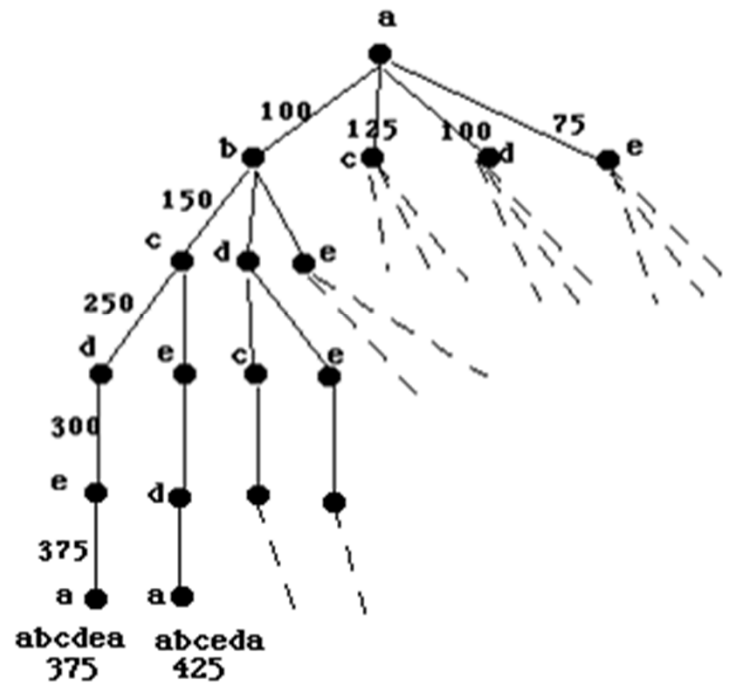
Search - problem representation

- State Space: State space graph

An Instance of the
Traveling Salesman Problem

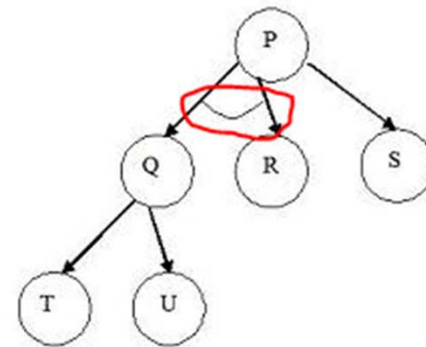


Search Space



Search - problem representation

- AND/OR Graph
 - Horizontally connected edges (here marked in red) represent AND nodes
 - For AND nodes, each of the nodes have to be solved
 - Eg
 - problem reduction
 - Games (eg chess)

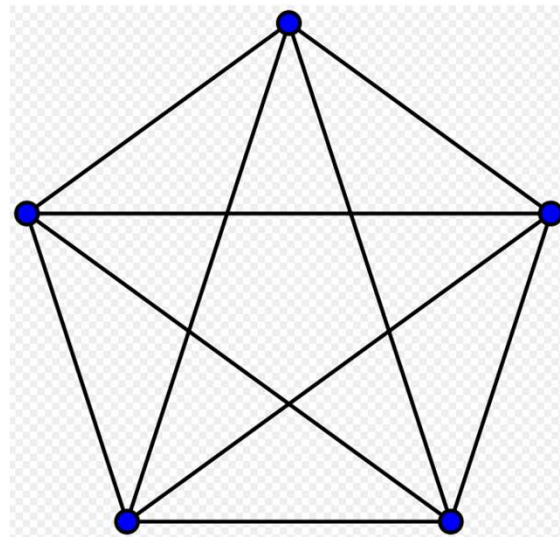


Search – Blind Search

- The Blind search algorithms following
 - Breadth First Search
 - Depth First Search
 - Uniform Cost Search
 - Assume the State Space graph is a Directed Tree
- The Heuristic search algorithms following
 - Ordered Search
 - A* An optimal search for an optimal solution
 - Assume the State Space graph is a General Graph

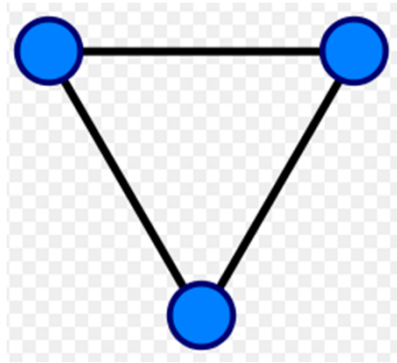
Search

- General graph consists of
 - Nodes or points
 - Arcs or edges connecting two nodes



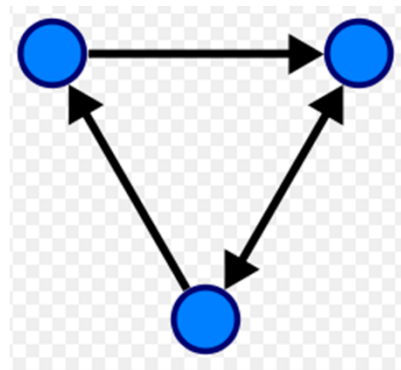
Search

- Arcs can be
 - Undirected



or

- Directed



Search

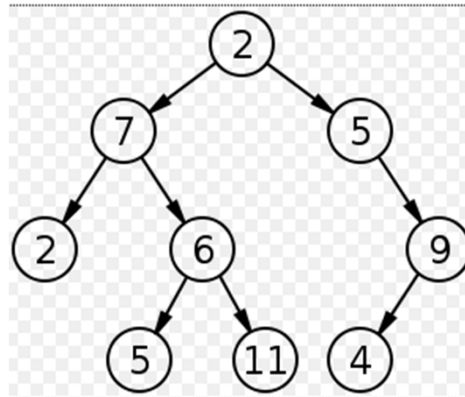
- Undirected graph
 - Contains only undirected arcs
- Directed graph or digraph
 - Contains only directed arcs
- Mixed graph
 - Contains both directed and undirected arcs

Search

- In a directed graph (containing only directed arcs)
- The indegree of a node
 - Is the number of arcs terminating in that node
- The outdegree of a node
 - Is the number of arcs starting in that node

Search

- A Directed tree
 - Is an acyclic digraph
 - Which has one node called the root
 - The root node has indegree zero, and
 - All other nodes have indegree one



Search – Blind State Space Search

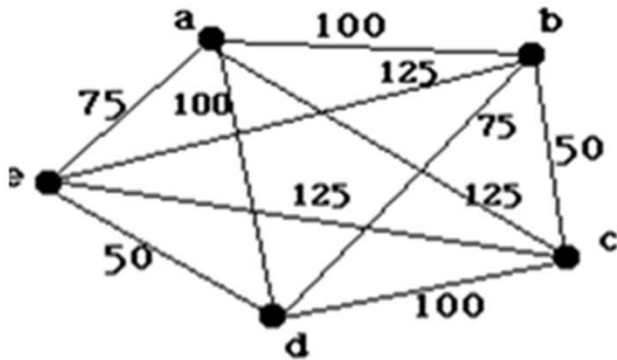
Breadth-First Search

- Expands nodes in their proximity from the root (or start) node
- Expands all nodes of depth n before expanding nodes of depth $n+1$
- Guaranteed to find the shortest possible solution

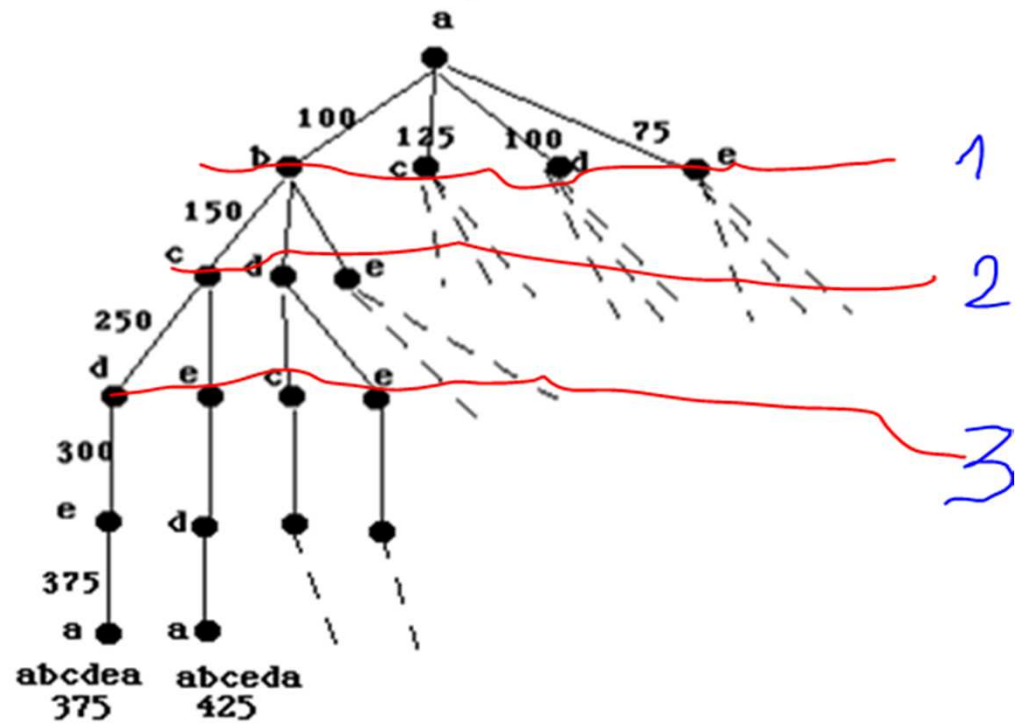
Search – Blind State Space Search

Breadth-First Search

An Instance of the Traveling Salesman Problem



Search Space



Search – Blind State Space Search

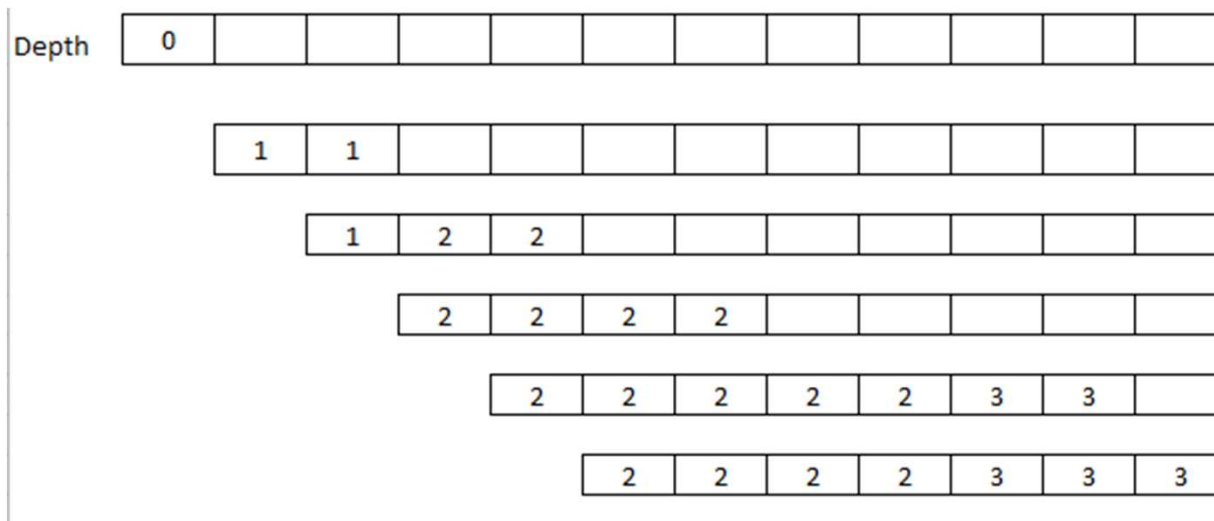
Breadth-First Search - Algorithm

1. Put the start node on a list, called OPEN, of unexpanded nodes
2. If OPEN is empty, no solution exists
3. Remove the first node, n , from OPEN and put it in a list, called CLOSED, of expanded nodes
4. Expand node n . If it has no successors, go to (2)
5. Place all successors of n at the **end** of the OPEN list
6. If any of the successors of n is a goal node, a solution has been found. Otherwise go to (2)

Search – Blind State Space Search

Breadth-First Search - Algorithm

Newly expanded nodes are added to the **end** of the list

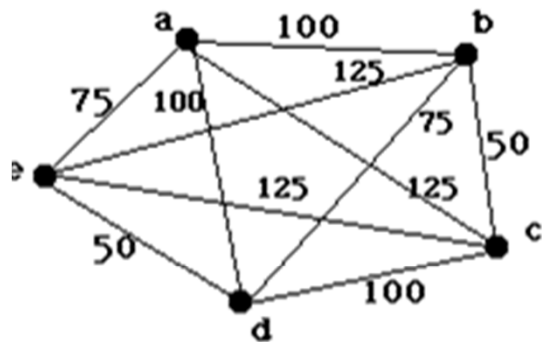


Search – Blind State Space Search

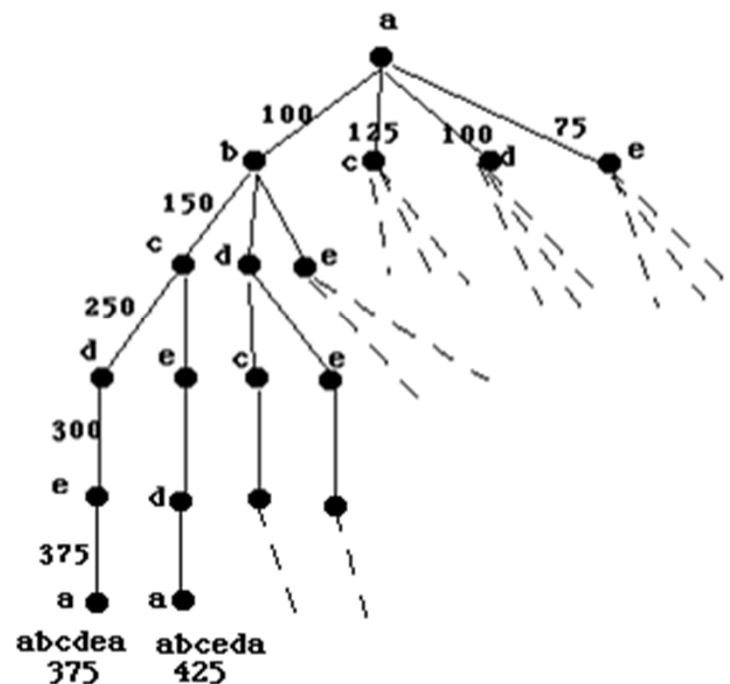
Depth-First Search

- Expands most recent (deepest) nodes first
 - Here abcdea first

An Instance of the Traveling Salesman Problem



Search Space



Search – Blind State Space Search

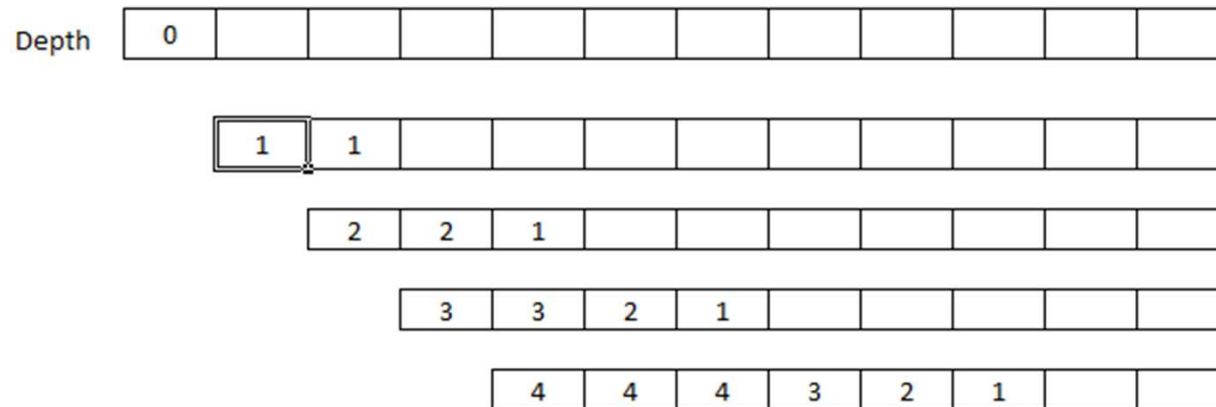
Depth-First Search - Algorithm

1. Put the start node on a list, called OPEN, of unexpanded nodes
2. If OPEN is empty, no solution exists
3. Remove the first node, n , from OPEN and put it in a list, called CLOSED, of expanded nodes
4. Expand node n . If it has no successors, go to (2)
5. *If the depth of node n is greater than the maximum depth, go to (2)*
6. Place all successors of n at the **beginning** of OPEN list
7. If any of the successors of n is a goal node, a solution has been found. Otherwise go to (2)

Search – Blind State Space Search

Depth-First Search

- Newly expanded nodes are added at the **beginning** of the list



Search – Blind State Space Search

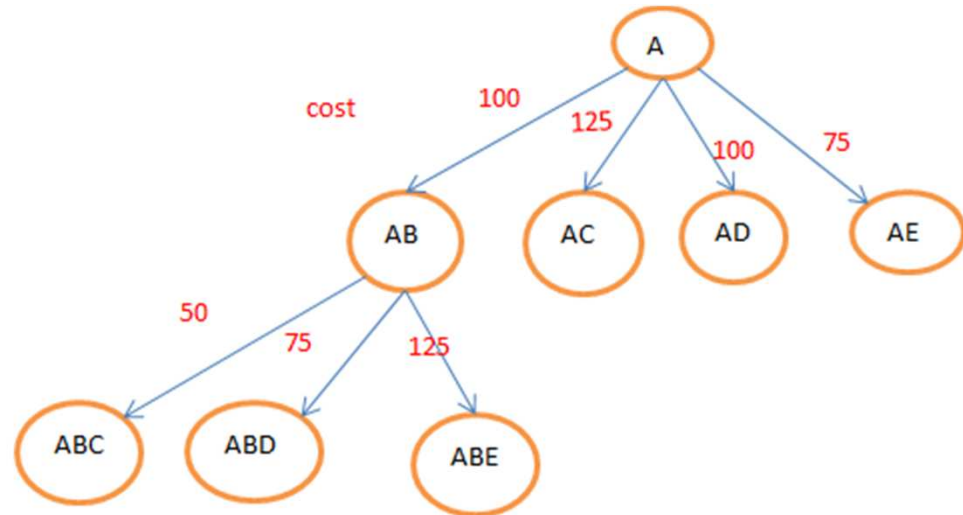
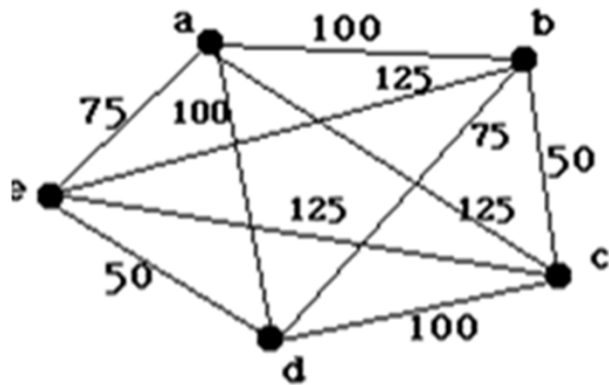
Uniform Cost Search

- The Breadth-First search can be generalized slightly to solve to problem of finding the cheapest path from a start node to a goal state
- A non-negative cost is associated with each arc joining two nodes
- The cost of a solution is then the sum of all the costs along the path

Search – Blind State Space Search

Uniform Cost Search

An Instance of the
Traveling Salesman Problem



Search – Blind State Space Search

Uniform Cost Search - Algorithm

1. Put the start node on a list, called OPEN, of unexpanded nodes
2. If OPEN is empty, no solution exists
3. Select from OPEN a node i such that $\text{TotalCost}(i)$ is minimum. If several nodes qualify choose i to be a goal node if there is one, otherwise choose among them arbitrarily.
4. Remove node i from OPEN and place it on a list CLOSED of expanded nodes
5. If node i is a goal node, a solution has been found
6. Expand node i , if it has no successors go to (2)
7. For each successor node j of i
 - Compute $\text{TotalCost}(j) = \text{TotalCost}(i) + \text{Cost}(j)$
 - Add node j to the OPEN list
8. Go to (2)

Search – Blind State Space Search

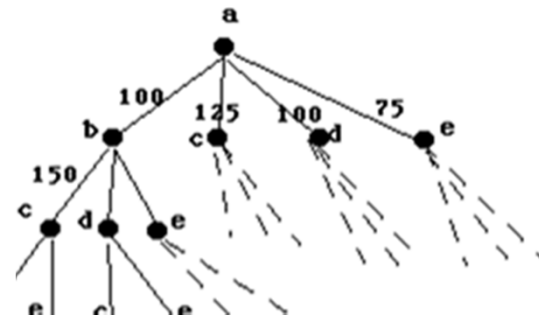
Uniform Cost Search

- If we associate the cost of node i to node j with
 - -1
 - the Uniform Cost Search becomes a Depth-First Search
 - Since TotalCost of the node = - Depth of the node
 - 1
 - the Uniform Cost Search becomes a Breadth-First Search
 - Since TotalCost of the node = Depth of the node

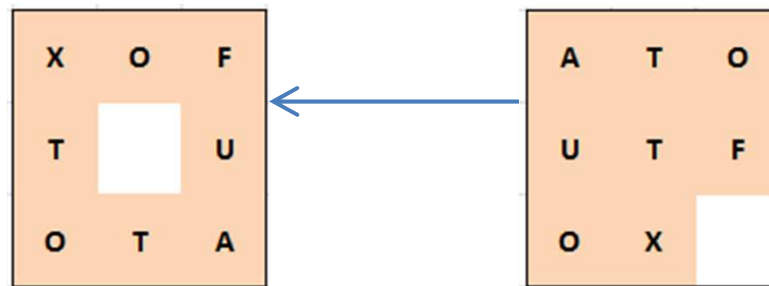
Search – Blind State Space Search

Bidirectional Search

- The algorithms so far use **forward reasoning**, ie moving from the start node towards a goal node



- In some cases we could use **backward reasoning**, ie moving from the goal state to the start state



Search – Blind State Space Search

Bidirectional Search

- Forward and backward reasoning can be combined into a technique called **bidirectional search**
- The idea is to replace a single search graph – which is likely to grow exponentially – by two smaller graphs
 - One starting from the initial state and searching forward
 - One starting from the goal state and searching backwardThe search terminates when the two graphs intersect

Search – Blind State Space Search

Bidirectional Search

- A bidirectional version of the Uniform Cost Search, guaranteed to find the shortest solution path, is due to Pohl (1969, 1971)
- Empirical data for randomly generated graphs expanded only $\frac{1}{4}$ as many nodes as unidirectional search

Search – Limiting the search

- The amount of time and space is critical to find a solution
 - Heuristics
 - Relaxing the requirement
 - Any (fast) solution, but not necessarily the best

Search - Heuristics

- In blind search the number of nodes can be extremely large
 - The order of expanding the nodes is arbitrary
 - Blind search does not use any properties of the problem being solved
 - Result is the combinatorial explosion
- Information about a particular problem can help to reduce the search
 - The question then is: how to search the given space efficiently

Search - Heuristics

- Heuristic information
 - Additional information beyond that which is built into the state and operator definitions
- Heuristic search
 - A search method using that heuristic information
 - Whether or not the method is foolproof
- Most of the programs were written for a single domain – heuristics were closely intertwined in the program and not accessible for study and adaptation to new problems

Search - Heuristics

- Heuristic search
 - Strategy to limit (drastically) the search for solutions in large problem spaces
- Ways of using heuristic information
 - Which node(s) to expand first instead of expanding in a strictly depth-first or breadth-first manner
 - When expanding a node, decide which successors to generate instead of blindly generating all successors at one time
 - Which nodes not to expand at all (pruning)

Search – Heuristics

Ordered or Best-Fit Search

- Addresses only the first point
 - Which node to expand first
 - Expands fully
 - The idea is to expand the node that seems **most promising**
 - The promise of a node can be defined in several ways
 - Estimate its distance to the goal node
 - Estimate the length of the entire path
 - In all cases the measure of promise of a node is estimated by calling an **evaluation function**

Search – Heuristics

Ordered or Best-Fit Search

- The basic algorithm is given by Nilsson (1971)
- The evaluation function f^* is defined so that the more promising a node is, the smaller is the value of f^*
 - Estimates its distance to the goal node
- The node selected for expansion is the one at which f^* is minimum
- The search space is assumed to be a general graph

Ordered or Best-Fit Search - Algorithm

1. Put the start node s on a list called OPEN of unexpanded nodes. Calculate $f^*(s)$ and associate its value with node s
2. If OPEN is empty, exit with failure; no solution exists
3. Select from OPEN a node i such that $f^*(i)$ is minimum. If several nodes qualify choose i to be a goal node if there is one, otherwise choose among them arbitrarily.
4. Remove node i from OPEN and place it on a list CLOSED of expanded nodes
5. If i is a goal node, exit with success; a solution has been found

(continued on next slide)

Ordered or Best-Fit Search - Algorithm

6. Expand node i
7. For each successor node j of i
 - a. Calculate $f^*(j)$
 - b. If j is neither in list OPEN or list CLOSED, add it to OPEN. Attach a pointer back from j to its predecessor i
 - c. If j was already on OPEN or CLOSED, compare the f^* value just calculated with the previously calculated value
 - d. If the new value is lower
 - i. Substitute it for the old value
 - ii. Point j back to i instead of to its previously found predecessor
 - iii. If j was on the CLOSED list, move it back to OPEN
8. Go to (2)

A* Optimal search for an optimal solution

- Ordered search looked only at the promise of the node, not necessarily at the minimum cost or path
- We can change f^* slightly to find a minimum cost solution
- $f^*(n) = g^*(n) + h^*(n)$
 - $g^*(n)$ estimates the minimum cost of the start node to n
 - $h^*(n)$ estimates the minimum cost of n to the goal
- h^*
 - is the carrier of heuristic information
 - Should never overestimate the cost
 - $h^*(n) \leq h(n)$

Example – Breadth-First Search

- Given a number of amounts, try to find a combination of amounts that matches another amount
- Example
 - We have 10, 20, 30, 40,, 100
 - Find all combinations that produce eg 120
- We use Breadth-First search, ie find first those with a minimum of depth – more likely

Total number of combinations

- If we have n amounts
- $\sum_{p=0}^n Comb(n, p)$
 - $Comb(n, p) = n! / ((n-p)! * p!)$
- Example with 50 amounts
 - $\sum_{p=0}^{50} Comb(50, p) = 1,125,899,906,842,623$

Heuristics

- Heuristic information
 - Sort amounts in ascending sequence
 - For each index
 - Calculate minimum sum ahead
 - Calculate maximum sum ahead
- Heuristic search
 - Prune as soon as possible

Index	1	2	3	4	5	6	7	8	9
Amounts	-30	-20	-10	40	50	60	70	80	90
Minimum sum ahead	-60	-30	-10	40	50	60	70	80	90
Maximum sum ahead	390	390	390	390	350	300	240	170	90
LastNegativeIndex	3								

Heuristics

- Construction of Min/Max sum ahead

Index	1	2	3	4	5	6	7	8	9
Amounts	-30	-20	-10	40	50	60	70	80	90
Minimum sum ahead	-60	-30	-10	40	50	60	70	80	90
Maximum sum ahead	390	390	390	390	350	300	240	170	90
LastNegativeIndex	3								
Minimum sum ahead	- copy Amounts								
	for i = LastNegativeIndex - 1 to 1								
	[i] = [i] + [i + 1]								
	endfor								
Maximum sum ahead	- copy Amounts								
	for i = maxIndex - 1 to LastNegativeIndex + 1								
	[i] = [i] + [i + 1]								
	endfor								
	for i = LastNegativeIndex to 1								
	[i] = [i + 1]								
	endfor								

Heuristics

Index	1	2	3	4	5	6	7	8	9		
Amounts	-30	-20	-10	40	50	60	70	80	90		
Minimum sum ahead	-60	-30	-10	40	50	60	70	80	90		
Maximum sum ahead	390	390	390	390	350	300	240	170	90		
LastNegativeIndex	3										
(1) Search for 500	500 is not between the min and max of the first index										
	We don't even have to start searching - pruned all 511 combinations										
(2) Search for 70 with 4 combinations						look for	index	Value	MinAhead	MaxAhead	between
	We have index 4, ie 40					70	4	40	40	390	yes
	Continue searching from index 5 onwards					30	5	50	50	350	no
	still need 70-40 or 30										
	30 is not between 50 and 350										
	prune : comb(9-5+1, 4-2) =10										

Heuristics

- Formula to calculate how many nodes are going to be pruned

Index	1	2	3	4	5	6	7	8	9
Amounts	-30	-20	-10	40	50	60	70	80	90
Minimum sum ahead	-60	-30	-10	40	50	60	70	80	90
Maximum sum ahead	390	390	390	390	350	300	240	170	90

General pruned formula

n = number of amounts in array

$maxDepth$ = maximum depth investigating

$currentDepth$ = current depth

$investigatingIndex$: index investigating for further expansion

$pruned = comb(n - investigatingIndex + 1, maxDepth - currentDepth)$

Example (2) of previous page

$n = 9$

$maxDepth = 4$

$currentDepth = 1$

$investigatingIndex = 5$

$comb(9 - 5 + 1, 4 - 1)$

Heuristics

- Sign Reversal : try to prune as early as possible
- Decrease the gaps between minimum sum ahead and maximum sum ahead as soon as possible
- Make sure the number with the highest absolute value is at the front



- Reverse the sign of all amounts + the amount to search if necessary

Heuristics

Index	1	2	3	4	5	6
Amounts	-30	-20	-10	1010	1020	1030
Minimum sum ahead	-60	-30	-10	1010	1020	1030
Maximum sum ahead	3060	3060	3060	3060	2050	1030
Gap Minimum/Maximum	3120	3090	3070	2050	1030	0
Gap Delta		30	20	1020	1020	1030
Biggest abs() in front						
Index	1	2	3	4	5	6
Amounts	-1030	-1020	-1010	10	20	30
Minimum sum ahead	-3060	-2030	-1010	10	20	30
Maximum sum ahead	60	60	60	60	50	30
Gap Minimum/Maximum	3120	2090	1070	50	30	0
Gap Delta		1030	1020	1020	20	30

Heuristics

Index	1	2	3	4	5	6	7	8	9	10
Amounts	1	2	3	4	5	6	7	8	9	10
Minimum sum ahead	1	2	3	4	5	6	7	8	9	10
Maximum sum ahead	55	54	52	49	45	40	34	27	19	10
Gap Minimum/Maximum	54	52	49	45	40	34	27	19	10	0
search	5	5	5	5	5	5	5	5	5	5
take	1	2	3	4	5	6	7	8	9	10
new find	4	3	2	1	0	-1	-2	-3	-4	-5
Index	1	2	3	4	5	6	7	8	9	10
Amounts	-10	-9	-8	-7	-6	-5	-4	-3	-2	-1
Minimum sum ahead	-55	-45	-36	-28	-21	-15	-10	-6	-3	-1
Maximum sum ahead	-1	-1	-1	-1	-1	-1	-1	-1	-1	-1
Gap Minimum/Maximum	54	44	35	27	20	14	9	5	2	0
search	-5	-5	-5	-5	-5	-5	-5	-5	-5	-5
take	-10	-9	-8	-7	-6	-5	-4	-3	-2	-1
new find	5	4	3	2	1	0	-1	-2	-3	-4